

Recall: last time:

Mg: \longrightarrow Fair Game. $E(X_t | \mathcal{F}_s) = X_s$

Checked: BM is a Mg: $E(W(t) | \mathcal{F}_s) = W(s)$.

Quadratic Variation:

$$[X, X](T) = \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} (X(t_{i+1}) - X(t_i))^2$$

\swarrow
 $\max(t_{i+1} - t_i)$

Compute $[W, W](T) = T$



Check: $W(t)^2 - [W, W]$ is a Mg.

Theorem 1: Let M be a cts mg with filt $\{\mathcal{F}_t\}$.

Then $E(M(t)^2) < \infty \iff E[M, M](t) < \infty$.

in this case $E M(t)^2 = E M(0)^2 + E [M, M](t)$

Moreover $M(t)^2 - [M, M](t)$ is a mg.

Theorem 2: If A is a cts ^{adapted} _{increasing} process with $A(0) = 0$

& M is a cts Mg.

Then If $M^2 - A$ is a mg $\implies A = [M, M]$.

Intuition: If X has finite Q.V. \Rightarrow
 $(X(t+\delta t) - X(t))^2 \approx O(\delta t)$

If X has finite first variation

Expect $(X(t+\delta t) - X(t)) \approx O(\delta t)$.

Construction of Ito Integral

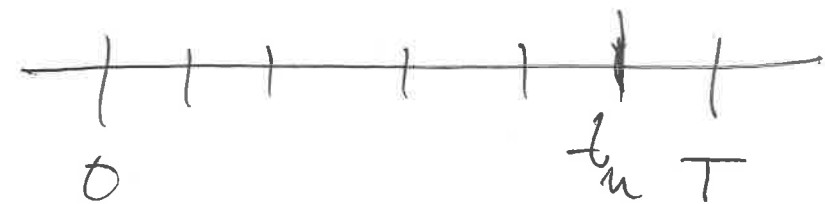
$W \rightarrow$ std BM, $\{\mathcal{F}_t\} \rightarrow$ Brownian filtration.

Let D be an adapted process.

($D(t)$ = your position on an asset price is $W(t)$).

$$\text{let } P = \{0 = t_0 < t_1 \dots < t_n \leq T\}.$$

Assume we only trade at times t_i



$$\text{let } I_P(T) = \sum_{i=0}^{n-1} D(t_i) (W(t_{i+1}) - W(t_i)) + D(t_n) (W(T) - W(t_n)).$$

whenever $t_n \leq T < t_{n+1}$.

lemma : ① $E I_P(T)^2 = E \left[\sum_{i=0}^{n-1} D(t_i)^2 (t_{i+1} - t_i) + D(t_n)^2 (T - t_n) \right]$
if $T \in [t_n, t_{n+1})$.

② I_P is a martingale.

③ $[I_P, I_P](T) = \sum_{i=0}^{n-1} D(t_i)^2 (t_{i+1} - t_i) + D(t_n)^2 (T - t_n)$
if $T \in [t_n, t_{n+1})$.

lets Check (1). (Assume $T = t_n$ for simplicity).

Notation: $\Delta_i W = W(t_{i+1}) - W(t_i)$.

$$I_p(T) = \sum D(t_i) \Delta_i W \quad \text{if } T = t_n.$$

$$\Rightarrow E(I_p(T))^2 = E\left(\sum_{i=0}^{n-1} D(t_i) \Delta_i W\right)^2.$$

$$= E \underbrace{\sum_{i=0}^{n-1} D(t_i)^2 (\Delta_i W)^2}_{(1)} + 2 \sum_{j=0}^{n-1} \sum_{i=0}^{j-1} E D(t_i) D(t_j) \underbrace{\Delta_i W \Delta_j W}_{\text{Claim} = 0}.$$

Claim = 0
→ (2)

Compute ①: $E \sum_{i=0}^{n-1} D(t_i)^2 (\Delta_i W)^2$

$$\approx \sum_{i=0}^{n-1} E \left(D(t_i)^2 (W(t_{i+1}) - W(t_i))^2 \right)$$

$$= \sum_{i=0}^{n-1} E E \left(D(t_i)^2 (W(t_{i+1}) - W(t_i))^2 \mid \mathcal{F}_{t_i} \right)$$

$$= \sum_{i=0}^{n-1} E \left[D(t_i)^2 E \left(\underbrace{W(t_{i+1}) - W(t_i)} \right)^2 \mid \mathcal{F}_{t_i} \right]$$

$$= \sum_{i=0}^{n-1} E D(t_i)^2 (t_{i+1} - t_i), \quad //$$

Check $E(z) = 0$: $i < j$

~~$m-1$~~
 ~~$j=0$~~

$$E(D(t_i) D(t_j) \Delta_i W \Delta_j W)$$

$$= E E(D(t_i) D(t_j) \Delta_j W \Delta_i W \mid \mathcal{F}_{t_j})$$

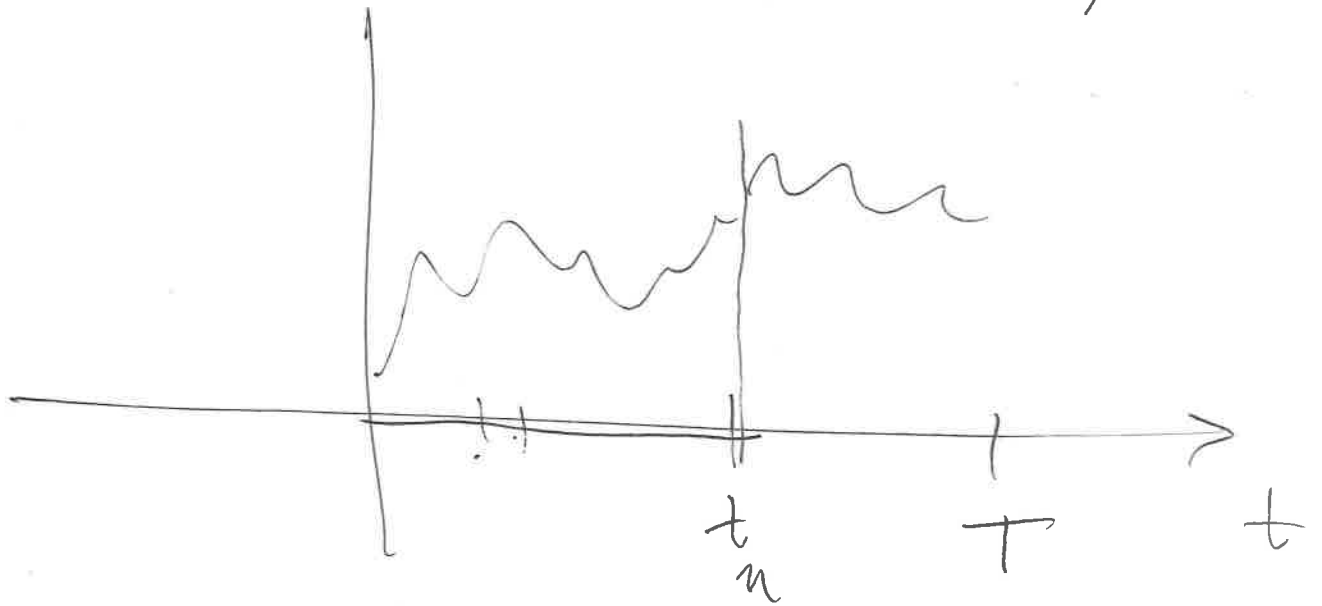
$$= E(D(t_i) D(t_j) \Delta_i W \underbrace{E(\Delta_j W \mid \mathcal{F}_{t_j})}_{0})$$

$$= 0$$

② You check I_P is a mg.

$$\textcircled{3} \quad I_P(T) = I_P(t_n) + D(t_n)(W(T) - W(t_n)).$$

$$\hookrightarrow [I_P, I_P](T) = [I_P, I_P](t_n) + D(t_n)^2 (T - t_n).$$



Ito's Idea:

Take partitions. γ $\|P\| \rightarrow 0$.

$$\text{Note } [I_P, I_P](T) = \sum_{i=0}^{n-1} D(t_i)^2 (t_{i+1} - t_i) \\ \uparrow + D(t_n)^2 (T - t_n).$$

Riemann Sum.

$$\lim_{\|P\| \rightarrow 0} [I_P, I_P](T) = \int_0^T D(t)^2 dt$$

Use this to show: the processes I_P themselves converge as $\|P\| \rightarrow 0$. The limit is the Ito integral.

Thm: $\mathbb{E} \int_0^T D(t)^2 dt < \infty$ almost surely.

The processes (I_φ) converge as $\|P\| \rightarrow 0$

$$\text{Let } \underline{I}(T) = \lim_{\|P\| \rightarrow 0} I_\varphi(T) = \int_0^T D(t) dW(t)$$

\hookrightarrow Itô integral.

Moreover $\mathbb{E} \int_0^T D(t)^2 dt < \infty$, then

$$I_\varphi \text{ is a martingale. } [I, I](T) = \int_0^T D(t)^2 dt$$

Remark: Key property needed is that D is adapted.

$$I_P(T) = \sum D(t_i) (W(t_{i+1}) - W(t_i)).$$

looks like Riemann sum.

Riemann Int: OK to look at

$$\lim_{\|P\| \rightarrow 0} \sum D(\xi_i) (W(t_{i+1}) - W(t_i))$$

where $\xi_i \in [t_i, t_{i+1}]$.

Will not work for Ito: In the above, to get the Ito int
need $\xi_i = t_i$.

Corollary (Ito Isometry).

If $E \int_0^T D(t)^2 dt < \infty$, then

$$E \left(\int_0^T D(t) dW(t) \right)^2 = E \int_0^T D(t)^2 dt.$$

Pf: I know $I(t) = \int_0^t D(t) dW(t)$ is a mg.

$$I(0) = 0$$

$$E I(T)^2 = E [I, I](T) = E \int_0^T D(t)^2 dt.$$

Goal: Ito's Formula (Generalization of Chain Rule).

Use \uparrow to compute.

Let b & σ be 2 adapted processes.

$$\text{Let } X(t) = X(0) + \underbrace{\int_0^t b(s) ds}_{\text{Riemann Int}} + \underbrace{\int_0^t \sigma(s) dW(s)}_{\text{Ito integral}}.$$

Ito process: ① $X(0)$ non random.

② $E \int_0^t \sigma(s)^2 ds < \infty$ & $\int_0^t b(s) ds < \infty$.

Short hand Notation for

$$X(t) = X(0) + \int_0^t b(s) ds + \int_0^t \sigma(s) dW(s).$$

It is

$$dX(t) = b(t) dt + \sigma(t) dW(t).$$

Q: What is $[X, X](t)$. $[X, X](T) = \int_0^T \sigma(s)^2 ds$.

Proof: $dX = b dt + \sigma dW$.

Then $[X, X](T) = \int_0^T \sigma(s)^2 ds$.

Proof: Let $B(t) = \int_0^t b(s) ds$. \leftarrow Bounded variation
(Finite first variation.)

$M(t) = \int_0^t \sigma(s) dW(s)$. \leftarrow Mg part.

$$X(T) = X(0) + B(T) + M(T)$$

$$\Delta_i X = X(t_{i+1}) - X(t_i), \quad \Delta_i B = B(t_{i+1}) - B(t_i)$$

$$\Delta_i M = M(t_{i+1}) - M(t_i).$$

$$\sum (\Delta_i X)^2 = \underbrace{\sum (\Delta_i M)^2}_{\text{converges to } [M, M]} + \underbrace{\sum (\Delta_i B)^2}_{\text{NTS} \rightarrow 0} + 2 \underbrace{\sum (\Delta_i M)(\Delta_i B)}_{\text{NTS} \rightarrow 0}$$

$$= \int_0^T \sigma(s)^2 ds.$$

Note. $(\Delta_i B)^2 = (B(t_{i+1}) - B(t_i))^2 = \left(\int_{t_i}^{t_{i+1}} b(s) ds \right)^2$.

$$\leq (\max b)^2 (t_{i+1} - t_i)^2.$$

$$\Rightarrow \sum (\Delta_i B)^2 \leq (\max b)^2 \sum (t_{i+1} - t_i)^2 \leq (\max b)^2 \max(t_{i+1} - t_i) \cdot \sum (t_{i+1} - t_i)$$

$$\leq T (\max t)^2 \|P\| \xrightarrow{\text{lim}} 0$$

You check: $\lim_{\|P\| \rightarrow 0} \sum (\Delta_i^M) (\Delta_i^B) \rightarrow 0.$

Decomposition of X as a $mg^{(M)}$ + a process of finite
 first variation (B) is called the Ito decomposition.
 (& is unique).

$$\text{If } dX = b dt + \sigma dW,$$

& D is some adapted process, then

we define $\int_0^T D(t) dX(t) = \int_0^T D(t) b(t) dt + \int_0^T D(t) \sigma(t) dW(t).$

Riemann Int

Ito integral.

Suppose $dX = b \cdot dt + \sigma dW$.

$$\left(X(T) - X(0) = \int_0^T b(t) dt + \int_0^T \sigma(t) dW(t) \right)$$

Let $f = f(t, x)$ some non-random fn.

Let $Y(t) = f(t, X)$.

Guess.

$$dY = d(f(t, X)) = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dX + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} d[X, X]$$